

# **SDC Localization Competition**

# Introduction

- The goal of this competition is to develop a localization module for estimating the poses of a self-driving car given a map.
- Datasets:
  1. NuScenes dataset
  2. ITRI dataset

# Challenges

- There will be 2 different level scenes(easy and medium) for the localization.
- We offer a basic [localization approach](#) with ICP scan matching as the baseline. You are free to use any approaches the do the localization.  
**If you don't have any idea about the localization, have a look at the document.**
- Your algorithm(s) need to have a better performance than our baseline or you won't get any credit.
- You also need to compare your performance with other teams to get more credit.

# Challenges

- **Test**

NCTU campus (Just for practice)

- **Easy**

ITRI campus (Available in 4/22)

- **Medium**

Urban from Nuscenes (Available in 4/29)

# Submission rules

- In easy and medium level, we provide two types data for you. The public data just for you to try and the private data is for evaluation.
- Public data:

The data **with** ground truth for a few seconds. You can test your algorithm(s) with these data via the [evaluated program](#) by yourself.

- Private data:

The data **without** ground truth for a few seconds. **We will evaluate the final ranking with the private data.** You can keep uploading your result from private data once a day until the deadline, We will evaluate and rank all the team.

# Results format

- The file of your result need to be **csv file** which named “{teamnumber\_case\_number}.csv”. For example: 2\_easy\_2.csv
- In the end of this competition, all teams need to upload the codes. We will check if the code can be compiled and executed.
- The result should contains coordinate data for **every LiDAR timestamp**. If the result lost any timestamp, your would get penalty score.

# Results format

- The result contains the pose data of each timestamp like:

(timestamp, x , y, z, yaw, pitch, roll).

- Your result should look like this:

```
1532142096.997409000,-2.25029,7.63406,-2.75526,2.54263,0.000165367,0.00918172
1532142097.097504000,-2.28618,7.66551,-2.81172,2.53572,-0.00224623,0.00971326
1532142097.197595000,-2.28754,7.67501,-2.84654,2.5352,0.00103778,0.00730458
1532142097.297654000,-2.29148,7.67083,-2.84117,2.53507,0.000734188,0.00713133
1532142097.397762000,-2.28858,7.6676,-2.83532,2.5353,0.000528738,0.00695795
1532142097.497863000,-2.28363,7.66673,-2.82788,2.5351,0.00107041,0.00676941
1532142097.597912000,-2.28314,7.66582,-2.83533,2.53516,0.00143774,0.00646181
1532142097.698001000,-2.28343,7.66176,-2.84524,2.53515,0.00110299,0.00693442
1532142097.798112000,-2.28397,7.66676,-2.83604,2.53491,0.000514449,0.00722409
1532142097.898310000,-2.28165,7.66595,-2.8234,2.53543,-0.00114634,0.00940711
1532142097.998263000,-2.2795,7.66323,-2.82058,2.5358,-0.00237217,0.00936604
```

# Evaluation metrics

Below we define the metrics for the our localization task. Our final score is a weighted sum of root-mean-square (RMS) of translation score and rotation score.

$$\begin{aligned} FinalScore = & 0.4 \times TranslationScore_{RMS} + 0.4 \times YawScore_{RMS} + \\ & 0.1 \times PitchScore_{RMS} + 0.1 \times RollScore_{RMS} \end{aligned}$$



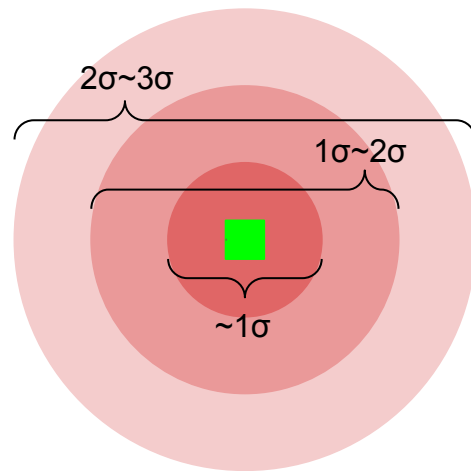
# Translation score RMS

We define the translation score RMS which is calculated by the L2 distance between your result and ground truth in three dimensional Euclidean space. The  $\sigma$  is the covariance of the ground truth.

Score for each scan:

- $\sim 1\sigma$ : 1
- $1\sigma \sim 2\sigma$ : 0.7
- $2\sigma \sim 3\sigma$ : 0.3
- $3\sigma \sim$ : 0

$$TranslationScore_{RMS} = \sqrt{\frac{\sum_{i=0}^n score_i^2}{n}}$$



# Rotation score RMS

- There are many ways of expression about the rotation like [Quaternion](#), [Matrix](#),....
- Here we define the score with [yaw, pitch, roll](#) form(radians) and compute them **independently**.

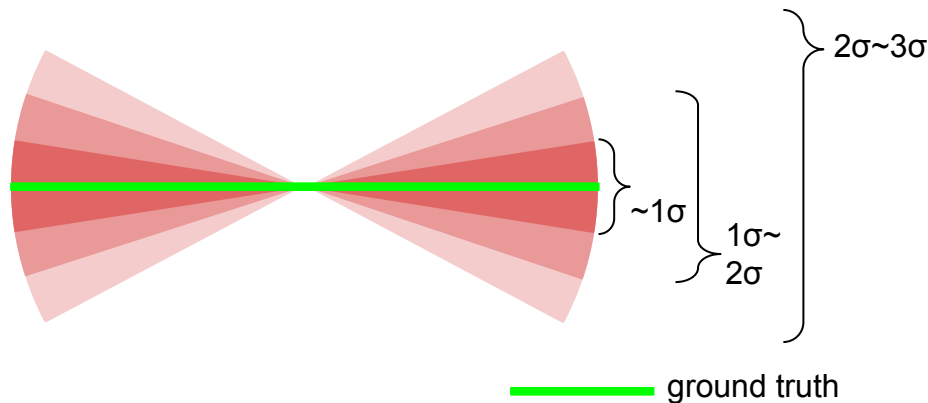
# Rotation score RMS

- We define the rotation score RMS which is calculated by the error between your result and ground truth in three dimensional Euclidean space. The  $\sigma$  is the covariance of the ground truth.

- Score for each scan:

- $\sim 1\sigma$ : 1
- $1\sigma \sim 2\sigma$ : 0.7
- $2\sigma \sim 3\sigma$ : 0.3
- $3\sigma \sim$ : 0

$$RotationScore_{RMS} = \sqrt{\frac{\sum_{i=0}^n score_i^2}{n}}$$



# Ranking and Grading

- Competition Ranking: 60%
- Presentation: 40%

# Competition Ranking

- Easy

- Top 10%: 50
- Top 25%: 45
- Top 50%: 40
- Others: 35
- Below baseline: 0

- Medium

- Top 10%: 50
- Top 25%: 45
- Top 50%: 40
- Others: 35
- Below baseline: 0

# Presentation

- Proposal: 10
- Report: 20
- Contribution: 30
- Final Presentation: 40

The contribution part is depends on the idea you implement or how you solve the issue you faced.

If you use whole open source project **without your idea**, you will get zero in the contribution part.

# Download

- [Dataset and evaluated program](#)